

APPENDIX A

	<u>Page</u>
A-1 NORMAL APPROXIMATION TO BINOMIAL	A-2
A-2 POISSON APPROXIMATION TO BINOMIAL	A-3
A-3 NORMAL APPROXIMATION TO POISSON	A-4

A-1. NORMAL APPROXIMATION TO BINOMIAL

When p , the probability of failure on a given trial, is moderate ($0.2 \leq p \leq 0.8$) and n the number of trials is large ($n \geq 30$), the normal distribution provides reasonable approximations to binomial probabilities. This approximation is detailed below. Note that Z is the notation for the standard normal variable. (See Appendix B, Table 2.)

The probability of k or fewer failures out of n trials is approximately equal to

$$1 - P(Z \geq (k + 0.5 - np) / \sqrt{np(1-p)}) .$$

The probability of at least k failures out of n trials is approximately equal to

$$P(Z \geq (k - 0.5 - np) / \sqrt{np(1-p)}) .$$

The probability of between k_1 and k_2 failures out of n trials inclusive is approximately equal to

$$P(Z \leq (k_1 - 0.5 - np) / \sqrt{np(1-p)}) - P(Z \geq (k_2 + 0.5 - np) / \sqrt{np(1-p)})$$

We have listed the approximations in the form

$$P(Z \geq a)$$

so that the use of Appendix B, Table 2 is direct.

As an example, suppose that $n = 40$ and $p = 0.3$. The probability of between 10 and 20 failures inclusive is

$$\begin{aligned} &P(Z \leq (10 - 0.5 - (40)(0.3) / \sqrt{(40)(0.3)(0.7)}) \\ &- P(Z \geq (20 + 0.5 - (40)(0.3) / \sqrt{(40)(0.3)(0.7)}) . \end{aligned}$$

Simplifying we obtain

$$P(Z \geq -0.86) - P(Z \geq 2.93) .$$

Now from Appendix B, Table 2, we find that $P(Z \geq -0.86) = 0.8051$ and $P(Z \geq 2.93) = 0.0017$. Consequently, the probability that between 10 and 20 failures inclusive occur is approximately 0.8034.

The value using a set of binomial tables is 0.8017.

A-2 . POISSON APPROXIMATION TO BINOMIAL

When p , the probability of failure on a given trial, is extreme ($p \leq 0.2$ or $p \geq 0.8$) and n , the number of trials, is large ($n \geq 30$), the **Poisson distribution** provides reasonable approximations to **binomial** probabilities. We make the identification $m = np$ and use Poisson tables to determine the probabilities of events in a binomial experiment.

As an example, suppose that $n = 40$ and $p = 0.05$, so that $m = 40(0.05) = 2$. The probability of between 5 and 10 failures is the difference between the probability of 10 or fewer failures (1.000) and the probability of 4 or fewer failures (0.947). (See Appendix B, Table 3.) The difference is 0.053. Using a set of binomial tables we obtain 0.0480.

A-3 . NORMAL APPROXIMATION TO POISSON

When the product λT is greater than 5, the normal distribution provides reasonable approximations to Poisson probabilities. The approximation is detailed below. Note that Z is the notation for the standard normal variable. (See Appendix B, Table 2)

The probability of k or fewer failures during time T is approximately

$$1 - P(Z \geq (k+0.5-\lambda T)/\sqrt{\lambda T}) .$$

The probability of at least k failures during time T is approximately

$$P(Z \geq (k-0.5-\lambda T)/\sqrt{\lambda T}) .$$

The probability of between k_1 and k_2 failures inclusive during time T is approximately

$$P(Z \geq (k_1-0.5-\lambda T)/\sqrt{\lambda T}) - P(Z \geq (k_2+0.5-\lambda T)/\sqrt{\lambda T}) .$$

We have listed the approximations in the form

$$P(Z \geq a)$$

so that the use of Appendix B, Table 2 is direct.

As an example, suppose that the failure rate λ is 0.01 and the test time T is 1000 hours. The probability of between 8 and 15 failures inclusive is

$$\begin{aligned} &P(Z \geq (8-0.5-(0.01)(1000))/\sqrt{(0.01)(1000)}) \\ &- P(Z \geq (15+0.5-(0.01)(1000))/\sqrt{(0.01)(1000)}) . \end{aligned}$$

The above expression reduces to

$$P(Z \geq -0.79) - P(Z \geq 1.74) .$$

Now $P(Z \geq -0.79) = 0.7852$ and $P(Z \geq 1.74) = 0.049$, so the **probability that** between 8-and 15 failures inclusive **occur** is approximately 0.7443.

Using the Poisson tables (Appendix B, Table 3), we obtain the probability more precisely as 0.731.